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Capacity Choice under Demand Uncertainty: Effects of Production Postponement and Product Flexibility

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Abstract: This paper deals with the optimal capacity choice under demand uncertainty. A single period two product model with stochastic demand has been developed to determine the optimal capacity level that maximizes the expected profit. Dedicated plant with no production postponement strategy has been considered as base case. The model has been extended to examine the effect of production postponement and product flexibility on optimal capacity decision. While it is apparent that the cost of over-production has been eliminated under production postponement, the other major benefit depends on whether the products have been produced in a single product flexible plant rather than dedicated plants. It has been shown that investment in flexible plant makes sense only if the possibility of production postponement exists. The model has been extended to multi-product situation with correlation in demand. Simulated data based optimization procedure has been applied to solve the multi-product problem as the same is analytically intractable. The concept of PdPPF Index has been introduced to observe the effect of production postponement on product flexible plant. Finally the effects of imposing service level objective on firm's optimal profit and capacity have been studied for both dedicated and flexible plant strategies.

Keywords: Demand Uncertainty; Capacity Planning; Production Postponement; Product Flexibility; Stochastic Programming; Service Level

1. Introduction:

Up to the middle of the last century, the paradigm of manufacturing had an emphasis on the mass production, mass markets and standard design. The existence of national market and absence of foreign competitors helped firms to act in the seller's market. Over the years the complexity in business environment has increased due to globalization and rapid technological advances. The changing nature of global business has led to highly competitive markets. Increased competition has changed the nature of demand in the market place both in terms of product variety as well as uncertainty associated with the product demands. This has increased challenges in all facets of manufacturing. Capacity planning in such scenario assumes complexity as one has to deal with the trade off between the cost of investment in excess capacity and the opportunity loss from not meeting the demand due to capacity constraint. In the context of production decision under capacity planning objective, two distinct situations may arise: (a) the firm has to decide on capacity as well as the production quantity before the demand has been realized, (b) while capacity needs to be decided a priori, the firm can decide on production after the demand is realized. The above two have been normally referred to as "No Postponement" and "Production Postponement" respectively. While it has been apparent that the cost of over-production has been eliminated under production postponement, the other major benefit depends on whether the products have been produced in a single product flexible plant rather than dedicated plants. Yang et al. (2004) have argued that both flexibility and postponement are "reactive adaption behaviors" as they deal with the consequences of uncertainty rather than attacking the causes of uncertainty.

Product flexibility has been recognized as an important tool for coping with demand uncertainties. However, investment and management issues regarding product flexibility have been recently incorporated in operation management models (Bish and Wang, 2004). It is intuitive that in the presence of production postponement, the firm stands to gain from product flexibility by exploiting the differences in the realized demands of the individual products. The capacity decision being taken considering aggregate demand of all the products; at the production stage potential benefit exists in terms of utilizing the idle capacity due to the below average realized demand for one product by the higher than average realized demand for another product. On the other hand, as product flexibility allows production of different products in the same plant, it would typically involve higher marginal cost of investment compared to dedicated plant. This motivates to look at the economics of dedicated plants versus product flexible plant in the context of capacity planning decision. For this purpose, a single period multi-product model with stochastic demand has been developed to determine the optimal capacity level that maximizes the expected profit. Dedicated plant with no production postponement strategy has been considered as base case. This model is similar to the classical newsboy model with capacity as decision variable. The model has been extended to consider (a) Dedicated plant with production postponement and (b) Flexible plant with production postponement. The base model as well as (a) above, are essentially extension of Mieghem and Dada (1999) for multi product case. In literature, the extensions (a) and (b) have been modeled as two stage stochastic programming problem; where, in the first stage the firm decides the capacity that maximizes the

expected profit. In the second stage, demands have been realized and the firm decides on production quantity.

In this paper the stochastic programming problem has been solved to determine the capacity level which maximizes the expected profit. However, in case of multiple products following correlated multivariate demand distribution, the problem becomes analytically intractable. Because of the analytical intractability, most of the literatures have come out with characterization of optimal solution with possibilities and dominant conditions. (Some of them have been discussed in literature review.) To make the problem analytically tractable, for three-product case, where demands follow correlated multivariate distribution, finite discretization of the random data allows writing the expectation in the form of summation and helps to solve the stochastic problem.

The rest of the paper is as follows. Relevant literature survey has been done in section 2. In section 3, the models for expected profit maximization and opportunity loss minimization under postponement and product flexibility have been introduced and shown that simulated data based optimization gives very good approximation of the analytical results in case of two-product example. In section 4, analysis has been done for multi product case. The results and insights of the analyses have also been shown in this section. In section 5, service level constraint has been added to observe its effect on various strategies. Section 6 concludes the paper.

2. Literature Survey:

The choice of dedicated and flexible plant combination for capacity planning under demand uncertainty has been considered by Fine and Freund (1990), followed by Meighem (1998) and Bish and Wang (2004). Eppen et al (1989) have considered capacity planning problem under risk and presented a mixed integer programming model based on a scenario planning approach. Peronne et al (2002) have tried to capture the economic advantage of flexible resource over the dedicated one.

Fine and Freund (1990) have worked with n different product families which can be produced in n dedicated plants or in a single flexible plant. K possible states of demand with known probability have been assumed. The market demand has been realized by the firm only after investment in capacity for the combination of dedicated and flexible plant. The capacity acquisition cost, revenue and production costs has been known. With the objective of profit maximization, followings have been included in their findings:

- a) There is no guarantee of getting a unique optimal solution if the total no of products is greater than or equal to three. This has been shown by a counter example.
- b) Shadow values at optimality for dedicated and flexible capacity constraint have been obtained. They have shown that, for a particular state of market demand, shadow value of flexible capacity is equal to the maximum of shadow values of dedicated capacities over all products. From that they have also derived the profitable condition for the investment in flexible capacity.
- c) With the increase in dedicated capacity cost, dedicated capacity decreases, flexible capacity increases and vice-versa. Decrease in any type of capacity cost profit increases.
- d) For downward sloping demand curves, in case of two products they have shown the relationship between the capacity, capacity cost and optimal profit for correlated demand scenario.

Mieghem (1998) has extended the works of Fine and Freund (1990) with same two product example with product one contribution is greater than that of product two. The benefits of product flexibility under uncertainty has been observed for the role of price and cost mix differentials in addition to demand correlation. He has expressed optimality condition in terms of dual variables. He has also highlighted the role of investment cost for choosing among possible investment strategies. For this he has defined two threshold values for flexible capacity cost. Similar to Fine and Freund (1990) he has observed substitution effect of marginal cost change on capacity. Similarly the investment in corresponding dedicated capacity has been increased with higher price. Increase in price differential increases flexible capacity and decreases dedicated capacity of less profitable product. He has proposed capacity investment strategy for perfectly positively correlated and perfectly negatively correlated demand under different conditions. Contradicting Fine and Freund (1990) he has shown that investment in flexible plant can give better benefit even in case of perfectly positively correlated demand if there is price difference between the products.

In line with the above literatures, Bish and Wang (2004) have also considered two-product case considering continuous distribution which makes it different from Fine and Freund (1990) and price dependent demand which makes it different from Mieghem (1998). The problem is two

stage stochastic programming in nature. They have divided demand space into six regions and for each region they have derived optimal closed form expression for optimal profit of stage two as a function of capacity vector. For stage one problem they have derived necessary and sufficient condition for optimal profit. They have also derived necessary and sufficient conditions for investment in flexible capacity. They have proposed capacity investment strategies under perfectly correlated demand conditions for different parameter values.

Eppen et al (1989) have considered a multiproduct, multiplant, multiperiod capacity planning problem. Three scenarios (or states of nature) have been specified for each year. They have argued that variance is not a good measure of risk in this environment and suggested an alternative based on expected downside risk. Their works have been based on following assumptions: 1) a retooling decision determines which products can be produced at a site as well as other cost and capacity parameters; 2) there is a changeover cost for shutting down a plant as well as for retooling it; 3) the demand has been realized before the production decision has been made and no inventory has been carried from period to period; 4) production levels can be altered within the time period in order to satisfy as closely as possible the demand that has been actually experienced; 5) the probability of a scenario occurring in a year is independent of earlier outcomes and the capacity of a plant depends upon the configuration chosen and at any period any plant should be under one and only configuration. Interest rate has been taken as 0.1. They have added a constraint for expected downside risk to the original problem of the form EDR (0) < 7.0, where 0 is the target value of desired profit. Expected profit and EDR has been calculated from histogram generated using 15 cases (3 scenarios and 5 periods).

Peronne et al (2002) have assumed the following for their theoretical model: 1) demands follow uniform distribution, 2) price depends on mean demand only and 3) the variable cost is same for both dedicated and flexible plants but investment costs are different. System wise profit has been maximized by maximizing each products profit. Investment cost, which has been expressed in terms of unit time multiplied by the service time of the product, in flexible plant has been depended on scope economy factors α and β . According to them, flexibility has been most effective when products with longest service times have been performed in most expensive dedicated resource. The cumulative scope economy factor α for flexible machine is (investment cost of flexible machine)/(investment cost of dedicated machine capable of producing the products that have been produced in flexible plant). This flexible investment cost is less than sum of the total dedicated cost and greater than any of the dedicated cost. Similarly, service time scope economy factor β for any product is (service time in flexible machine)/(total service time in dedicated machine capable of producing the products that have produced in flexible plant). The difference between flexible resource and dedicated resource has been presented in the form of hyperbola i.e. $\alpha\beta$ = constant.

By simulating truncated normal distribution in 10-product-10-plant case Jordan and Graves (1995) have shown that limited flexibility with single chain captures more than 90% of the benefit of total flexibility in terms of expected sales and capacity utilization. According to them, benefit of flexibility has been affected by two factors; demand correlation and total capacity relative to expected total demand. Negatively correlated products have been required to be in the same chain, but might not be in the same plant. If total capacity deviates far from the total expected demand, flexibility has no value. They have argued that there can not be single optimal plan; rather many near optimal plans exist. The product-plant links have been added based on the following rules: 1) try to equalize the no of plants (measure in total units of capacity) to which each product in the chain has been directly connected, 2) try to equalize the no of products (measure in total units of expected demand) to which each plant in the chain has been directly connected and 3) create circuit that encompasses as many plants and products as possible.

Fine and Freund (1990), Meighem (1998) and Bish and Wang (2004) have examined two product situation and analytically studied characteristics of the optimal solution for two product case. In contrast to them, simulated data based models, developed in this paper, have been capable of finding optimal profit and capacity under given parameter values for multiproduct case with complete characterization of demand correlation into the model. However separate values for marginal cost of capacity for dedicated and flexible plants have not been considered. Also, partial product flexibility discussed by Jordan and Graves (1995), has been considered as out of scope for this paper.

3. Two Product Cases:

Consider a manufacturer producing two products wants to set capacity level(s) before realizing the demands. Also consider that, after demand realization, there is no inventory carry over or backorder which can affect next period's planning. Remaining inventory has been sold at discount, called salvage value. Take D_i as demand and d_i as the realized demand for the product i. The price of product i is P_i per unit, cost is C_i per unit and salvage value is S_i per unit. The firm can decide the production quantity Q_i before demand for product i has been realized or, it can wait till demand realization so that no over-production happens. Similarly firm can go for two dedicated plants with capacity K_i or single flexible plant to produce the products with capacity K. In the following subsections the possible cases has been discussed. For simplicity, subscript i have been omitted in case of dedicated plant strategies.

Assume that the manufacturer starts with no initial resource(s) and incurs investment cost C(K). For simplicity, also assume that C(K) is linear in K, i.e., $C(K) = C_K K$, where C_K depends on whether the firm is using dedicated technology or flexible technology. It has been considered that same amount of capacity has been required to produce one unit of each product, so capacity has been expressed as the number of product units that can be produced. Moreover, there is no cost associated with producing away from installed capacity. These types of assumptions are common in literature.

3.1. Analytical Findings for Two Products:

3.1.1. Dedicated Plant, No Production Postponement:

As there is no production postponement the firm needs to decide both the capacity and production before the demand realization. So there is no point in invest in capacity higher than the production level. In other words, in case of no production postponement K = Q

Possible two situations have been described below with the help of under production and over production costs:

Situation	Profit	Opportunity loss
D > K	$(P-C-C_K)K$	$(P-C-C_K)(D-K)$
$D \leq K$	$[(P-C-C_K) - (S-C-C_K)]D + (S-C-C_K)K$	$(S - C - C_K)(D - K)$

So expected profit = $E(\Pi)$

$$= \int_{K}^{\infty} \left[\left(P - C - C_{K} \right) K \right] f(d) dd + \int_{0}^{K} \left[\left[\left(P - C - C_{K} \right) - \left(S - C - C_{K} \right) \right] dd + \left(S - C - C_{K} \right) K \right] f(d) dd$$

= $\left(P - C - C_{K} \right) K - \left(P - S \right) \int_{0}^{K} \left[\left(K - d \right) \right] f(d) dd$ (1)

Similarly expected opportunity loss = E(O)

$$= \int_{K}^{\infty} \left[\left(P - C - C_{K} \right) (d - K) \right] f(d) dd + \int_{0}^{K} \left[\left(S - C - C_{K} \right) (D - K) \right] f(d) dd$$

Now,
$$\frac{d}{dK} \int_0^K df(d) dd = Kf(K)$$

 $\frac{\partial E(\Pi)}{\partial K} = (P - C - C_K) - (P - S)[F(K) + Kf(K) - Kf(K)] = (P - C - C_K) - (P - S)F(K)$
Hence, $F(K) = \frac{(P - C - C_K)}{(P - S)}$ (2)

3.1.2. Dedicated Plant, Production Postponement:

In case of production postponement, production has been done only after demand realization. Hence there is no over production cost and Q = Min (D, K). However, there has been a need to consider overcapacity cost in this case. There can be two situations as described below:

Situation	Profit	Opportunity loss
D > K	$(P-C-C_K)K$	$(P-C-C_K)(D-K)$
$D \leq K$	$(P-C-C_K)D-C_K(K-D)$	$C_{K}(K-D)$

So expected profit

$$= E(\Pi) = \int_{K}^{\infty} [(P - C - C_{K})K] f(d) dd + \int_{0}^{K} [(P - C - C_{K})d - C_{K}(K - D)] f(d) dd$$

= $(P - C - C_{K})K - (P - C) \int_{0}^{K} [(K - d)] f(d) dd$ (3)
Similarly expected opportunity loss

$$= E(O) = \int_{K}^{\infty} \left[\left(P - C - C_{K} \right) (d - K) \right] f(d) dd - \int_{0}^{K} \left[C_{K}(K - D) \right] f(d) dd$$

Hence, F(K) = $\frac{(P - C - C_{K})}{(P - C)}$ (4)

For the strategies discussed above, following propositions have been developed.

Proposition 1: Production postponement always gives higher optimal capacity and profit for dedicated plant.

Proof: As C > S, (P - C) < (P - S). Hence, F(K) in eq. (4) > F(K) in eq. (2), where, F(.) is c.d.f. of the distribution and F(.) increases monotonically in K.

Again (P - C) < (P - S) implies $E(\Pi)$ in eq. (3) > $E(\Pi)$ in eq. (1).

Proposition 2: For normally distributed demand, in absence of production postponement, optimal capacity increases with the increase in variance as long as $\frac{(P-C-C_K)}{(P-S)} \ge 0.5$, else optimal capacity decreases with the increase in variance.

Proof: Consider demand follows normal distribution with mean μ and standard deviation σ . Then, $F(K) = \Phi(\frac{K-\mu}{\sigma})$.

From eq. (2), $K = \mu + \sigma \Phi^{-1} \left[\frac{(P - C - C_K)}{(P - S)} \right]$.

For the rest of this sub-section proofs have been done taking $\frac{(P-C-C_K)}{(P-C)} \ge 0.5$ for both the products.

Proposition 3: For normally distributed demand, optimal profit decreases with the increase in variance.

Proof:
$$\int_0^K \left[(K-d) \right] f(d) dd = \int_0^K \left[(K-d) \right] \frac{1}{\sigma\sqrt{2\Pi}} e^{\frac{1}{2} \left(\frac{d-\mu}{\sigma}\right)^2} dd = (K-\mu)F(K) + \sigma^2 \{ f(K) - f(0) \}.$$

With increase in σ this part increases, which in turn reduces $E(\Pi)$ in eq. (1) and eq. (3).

3.1.3. Product flexible Plant, No Production Postponement:

When there is no production postponement, it has been shown that there is no added benefit from being product flexible. On the other hand, the investment required might be more for achieving product flexibility. Take total capacity = K, where $K = Q_1 + Q_2$. Below possible situations and the profit values corresponding to those situations have been presented.

Situation	Profit
$D_1 > Q_1, D_2 > K - Q_1$	$(P_1 - C_1 - C_K)Q_1 + (P_2 - C_2 - C_K)(K - Q_1)$
$D_1 > Q_1, D_2 \le K - Q_1$	$(P_1 - C_1 - C_K)Q_1 + (P_2 - C_2 - C_K)D_2 + (S_2 - C_2 - C_K)(K - Q_1 - D_2)$
$D_1 \le Q_1, D_2 > K - Q_1$	$(P_1 - C_1 - C_K)D_1 + (S_1 - C_1 - C_K)(Q_1 - D_1) + (P_2 - C_2 - C_K)(K - Q_1)$
$D_1 \le Q_1, D_2 \le K - Q_1$	$(P_1 - C_1 - C_K)D_1 + (S_1 - C_1 - C_K)(Q_1 - D_1) + (P_2 - C_2 - C_K)D_2 + (S_2 - C_2 - C_K)(K - Q_1 - D_2)$

Proposition 4: For independent demands, flexibility does not generate any extra profit compared to corresponding dedicated plants, as long as productions have not been postponed.

Proof: Considering demands are independent, i.e. $f(d_1,d_2) = f(d_1)f(d_2)$, the expression for expected profit works out as,

$$\begin{split} E(\Pi) &= \left(P_1 - C_1 - C_K\right)Q_1 + \left(P_2 - C_2 - C_K\right)(K - Q_1) \\ &- \left(P_1 - S_1\right)\int_0^{Q_1} \left[Q_1 - d_1\right] f(d_1)dd_1 - \left(P_2 - S_2\right)\int_0^{K - Q_1} \left[K - Q_1 - d_2\right] f(d_2)dd_2 \\ Now, \frac{\partial E(\Pi)}{\partial K} &= \left(P_2 - C_2 - C_K\right) - \left(P_2 - S_2\right)F_2(K - Q_1) = 0. \\ Or, F_2(K - Q_1) &= \frac{\left(P_2 - C_2 - C_K\right)}{\left(P_2 - S_2\right)} = F_2(Q_2). \\ Similarly, \frac{\partial E(\Pi)}{\partial Q_1} &= \\ \left(P_1 - C_1 - C_K\right) - \left(P_2 - C_2 - C_K\right) - \left(P_2 - S_2\right)F_2(Q_1) - \left(P_2 - S_2\right)F_2(K - Q_1)\frac{\partial(K - Q_1)}{\partial Q_1} = 0. \\ Or, F_1(Q_1) &= \frac{\left(P_1 - C_1 - C_K\right)}{\left(P_1 - S_1\right)}. \end{split}$$

3.1.4. Product Flexible Plant:

Take, total capacity = K, where $Q_1 + Q_2 \le K$

Without the loss of generality, it has also been considered product 1 gives more contribution, i.e. $(P_1-C_1) \ge (P_2-C_2)$. Hence the firm will always try to meet the demand of product 1 first and after that it will go for product 2.

Possible situations and the profit and opportunity cost values corresponding to those situations are:

Situation	Profit	Opportunity loss
$D_1 + D_2 > K$	$(P_1 - C_1 - C_K)D_1 + (P_2 - C_2 - C_K)(K - D_1)$	$(P_2 - C_2 - C_K)(D_1 + D_2 - K)$
$D_1 + D_2 \leq K$	$(P_1 - C_1 - C_K)D_1 + (P_2 - C_2 - C_K)D_2 - C_K(K - D_1 - D_2)$	$C_{K}(K - D_{1} - D_{2})$

 $E(\Pi) = \int_0^\infty \int_{K-d_1}^\infty \left[(P_1 - C_1) d_1 + (P_2 - C_2) (K - d_1) \right] f(d_1, d_2) dd_1 dd_2$

+
$$\int_0^\infty \int_0^{K-d_1} [(P_1-C_1)d_1+(P_2-C_2)d_2]f(d_1,d_2)dd_1dd_2-C_KK$$

Again considering demands are independent, i.e. $f(d_1,d_2) = f(d_1)f(d_2)$, the expression for expected profit works out as,

$$E(\Pi) = [(P_1 - C_1) - (P_2 - C_2)]\mu_1$$

+ $(P_2 - C_2 - C_K)K - (P_2 - C_2)\int_0^\infty [\int_0^{K-d_1} [K - d_1 - d_2]f(d_2)dd_2]f(d_1)dd_1 \dots (6)$

Where μ_1 represents mean demand for product 1.

Proposition 5: For independent and normally distributed demands having negligible probability of having demand less than zero

a) Flexible plant optimal capacity is less than corresponding dedicated plant total capacities.

b) With the increase in demand variance, optimal capacity of the flexible plant increases, but the increase in optimal capacity is less than corresponding total increase in dedicated plant optimal capacity.

Proof: Consider demand for product i follows normal distribution with mean μ_i and standard deviation σ_i .

Now,
$$\int_0^\infty [F_2(K - d_1)] f_1(d_1) dd_1 \cong \int_{-\infty}^\infty [F_2(K - d_1)] f_1(d_1) dd_1 = \operatorname{Prob}(D_1 + D_2 \le K) = \Phi(\frac{K - \mu_1 - \mu_2}{\sqrt{\sigma_1^2 + \sigma_2^2}})$$

From eq. (7),
$$K = \mu_1 + \mu_2 + \sqrt{\sigma_1^2 + \sigma_2^2} \left\{ \Phi^{-1} \left[\frac{(P_2 - C_2 - C_K)}{(P_2 - C_2)} \right] \right\}$$

For dedicated plants, total capacity $= K_D = \mu_1 + \sigma_1 \Phi^{-1} \left[\frac{(P_1 - C_1 - C_K)}{(P_1 - C_1)} \right] + \mu_2 + \sigma_2 \Phi^{-1} \left[\frac{(P_2 - C_2 - C_K)}{(P_2 - C_2)} \right]$
 $K_D - K = \sigma_1 \Phi^{-1} \left[\frac{(P_1 - C_1 - C_K)}{(P_1 - C_1)} \right] + \sigma_2 \Phi^{-1} \left[\frac{(P_2 - C_2 - C_K)}{(P_2 - C_2)} \right] - \sqrt{\sigma_1^2 + \sigma_2^2} \left\{ \Phi^{-1} \left[\frac{(P_2 - C_2 - C_K)}{(P_2 - C_2)} \right] \right\}$
As $(P_1 - C_1) \ge (P_2 - C_2), \Phi^{-1} \left[\frac{(P_1 - C_1 - C_K)}{(P_1 - C_1)} \right] \ge \Phi^{-1} \left[\frac{(P_1 - C_1 - C_K)}{(P_1 - C_1)} \right]$
Hence, $K_D - K \ge \left[\sigma_1 + \sigma_2 - \sqrt{\sigma_1^2 + \sigma_2^2} \right] \Phi^{-1} \left[\frac{(P_1 - C_1 - C_K)}{(P_1 - C_1)} \right]$

As $\sigma_1 + \sigma_2 \ge \sqrt{\sigma_1^2 + \sigma_2^2}$, $K_D \ge K$. This proves the first part.

With the increase in σ_i , K_D and K both increases, but the increase in $\sigma_1 + \sigma_2$ is more compared to $\sqrt{\sigma_1^2 + \sigma_2^2}$. This proves the second part.

Although intuitive, however if one wants to establish the following, analytical difficulty happens in case of flexible plant profit.

For independent and normally distributed demands having negligible probability of having demand less than zero

a) Flexible plant optimal profit is more than corresponding dedicated plant total profits.

b) With the increase in demand variance, optimal profit of the flexible plant decreases, but the decrease in optimal profit is less than corresponding total decrease in dedicated plant optimal profit.

For this purpose one needs to show: $E(\Pi) \ge E(\Pi_D)$.

Where, from eq. (6),

$$E(\Pi) = \left[\left(P_1 - C_1 \right) - \left(P_2 - C_2 \right) \right] \mu_1 + \left(P_2 - C_2 - C_K \right) K$$
$$- \left(P_2 - C_2 \right) \int_0^\infty \left[\int_0^{K - d_1} \left[K - d_1 - d_2 \right] f(d_2) dd_2 \right] f(d_1) dd_1$$

From eq. (3), total profit for dedicated plants

$$= E(\Pi_D) = (P_1 - C_1 - C_K)K_1 - (P_1 - C_1)\int_0^{K_1} [(K_1 - d_1)]f(d_1)dd_1$$
$$+ (P_2 - C_2 - C_K)K_2 - (P_2 - C_2)\int_0^{K_2} [(K_2 - d_2)]f(d_2)dd_2$$

The derivation of flexible plant profit, $E(\Pi)$ has not been tried.

3.2. Simulated Data Based Optimization Procedure:

3.2.1. Methodology:

In the last section it has been observed that even for two product case with demands following independent distribution, finding a closed form solution for optimal profit and corresponding capacity is extremely difficult. The complexity increases if the demands are not independent. Only in some specific cases analytical calculation of stochastic programming is possible as the evaluation of expected value of demand involves calculation of multivariate integrals. A finite discretization of the random data allows writing the expectation in the form of summation and helps to solve the stochastic problem. In this sub-section this methodology has been developed. The model of flexible plant has been considered for this purpose.

The model for flexible plant:

 $\Pi_{Flexible \ Plant} = \sum_{i=1}^{m} [(P_i - C_i) Min(D_i, \ K_i)] - C_{Flexible \ Plant} K$

Take, $Z_i = Min(D_i, K_i) = Production$ quantity of product i, where, $\sum_{i=1}^{m} K_i = K$.

The deterministic version of the flexible plant model (where d_i values are known with certainty) can be written as:

$$\begin{array}{ll} \text{Maximize} & \sum_{i=1}^{m} \left[(P_i - C_i) Z_i \right] - C_{\text{Flexible Plant}} K \\ \text{Subject to:} & Z_i \leq D_i & \forall \ i \\ & \sum_{i=1}^{m} Z_i - K \leq 0 \\ & Z_i \in K \geq 0 & \forall \ i \end{array}$$

But in real life d_i values are not known. One has the idea of the distribution of the d_i values only and before the realization of these values one need to set the capacity K. To summarize this, time sequence is as follows (Wagner, 1993, Ch. 16, p. 667);

- a) First stage: Manufacturer selects level of K.
- b) Random event: Values of d_i are known and are independent of K.
- c) *Second stage*: Manufacturer selects the level of Z_i, the production quantity.

Given the time sequence, manufacturer selects K for which expected profit has been maximized. The problem can be formulated as stochastic programming with recourse in the following way: First stage problem:

 $\max_{K \ge 0} \Pi(K) = E[\Pi^*(K, \mathbf{D})] - C_{\text{Flexible Plant}} K$

Second stage problem:

 $\Pi^{*}(K, \mathbf{d}) = \max_{\mathbf{Z} \ge 0} \Pi(K, \mathbf{d}) = \max \sum_{i=1}^{m} [(P_{i} - C_{i})Z_{i}]$ Subject to: $Z_{i} \le d_{i} \quad \forall i$ $\sum_{i=1}^{m} Z_{i} - K \le 0$

Here, $\mathbf{D} = (D_1, D_2, \dots, D_m)$, $\mathbf{d} = (d_1, d_2, \dots, d_m)$ and $\mathbf{Z} = (Z_1, Z_2, \dots, Z_m)$. Also E(.) is the expectation operator.

Now, consider only three values of **d** are possible with known probability $p_i s$ where $\sum_{j=1}^{3} p_j = 1$.

Hence the possible cases are (considering two product case): d_{11} , d_{12} with probability p_1 ; d_{21} , d_{22} with probability p_2 and d_{31} , d_{32} with probability p_3 . Since K has been determined before the realization of the demand values, these variables will also come in stochastic programming formulation. The remaining decision variables Z_i have been determined after the realization of the demand. Hence, they have been noted as Z_{ij} for i = 1, 2 and j = 1, 2, 3.

As long as the decision variables in the first stage (here capacity K) do not depend on the realization of the random event, the two stage problem can be expressed as a single optimization model like below:

$$\begin{array}{lll} \mbox{Maximize} & \sum_{j=1}^{3} \bigl\{ \sum_{i=1}^{2} \bigl[(P_i - C_i) Z_{ij} \bigr] \bigr\} p_j - C_{Flexible \ Plant} \ K \\ \mbox{Subject to:} & Z_{ij} \leq d_{ij} & \forall \ i, \ j \\ & \sum_{i=1}^{2} Z_{ij} - K \leq 0 & \forall \ j \\ & Z_{ij}, \ K \geq 0 & \forall \ i, \ j \end{array}$$

Here, the stochastic programming version of the problem has more number of constraints compared to its deterministic version.

As the first stage variable K do not depend on the outcome of the j^{th} scenario, objective function can be rewritten as:

Maximize
$$-C_{\text{Flexible Plant}} K + E\left[\left\{\sum_{i=1}^{2}\left[(P_i - C_i)Z_{ij}\right]\right\}p_j\right]$$

Finally, the distribution of **d** has been approximated by taking large number of values generated from the distribution. So all p_i values are equally likely and the model becomes:

 $\begin{array}{ll} \text{Maximize} & & \frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{2} \left[(P_i - C_i) Z_{ij} \right] - C_{\text{Flexible Plant}} K \\ \text{Subject to:} & & Z_{ij} \leq d_{ij} & \forall \ i, \ j \\ & & \sum_{i=1}^{2} Z_{ij} - K \leq 0 & \forall \ j \\ & & Z_{ij}, \ K \geq 0 & \forall \ i, \ j \end{array}$

However, with this procedure one trade off has been necessary. On one side, with the increase in number of products sample sets needs to be increased, otherwise the gap between sample statistic and parameter value increases. On the other side, with the increase in sample values the complexity of the problem increases and with the increase in number of products the complexity increases exponentially. Hence, to keep the accuracy of the results high, too many products have not been considered.

As discussed earlier in proposition 4, flexible plant without production postponement is not better option compared to multiple dedicated plants. So this strategy has been omitted in this section. The additional notations used in the models have been shown below:

 d_{ij} = Demand for product i at iteration j

 $C_{\text{Strategy S}} = \text{marginal cost of capacity for strategy S}$

3.2.2. Models:

Two Product, Dedicated Plants, No Production Postponement:

 $\Pi_{\text{Strategy 1}} = \sum_{i=1}^{2} [P_i \text{Min}(D_i, K_i) + S_i \text{Max}(K_i - D_i, 0) - C_i K_i] - C_{\text{Strategy 1}} \sum_{i=1}^{2} K_i$ Take, $Z_i = \text{Min}(D_i, K_i) = \text{Sales quantity of product i in the primary market}$ Then, $\text{Max}(K_i - D_i, 0) = -\text{Min}(D_i - K_i, 0) = -\text{Min}(D_i, K_i) + K_i = -Z_i + K_i$ Hence the simulated data based optimization model becomes,

$$\begin{split} \text{Maximize} & \quad \frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{2} \left[(P_i - S_i) Z_{ij} \right] - \sum_{i=1}^{2} (C_{\text{Strategy } 1} + C_i - S_i) K_i \\ \text{Subject to:} & \quad Z_{ij} \leq d_{ij} \quad \forall \ i, j \\ & \quad Z_{ij} - K_i \leq 0 \quad \forall \ i, j \\ & \quad Z_{ij}, K_i \geq 0 \quad \forall \ i, j \end{split}$$

Two Product, Dedicated Plants, Production Postponement:

$$\begin{split} \Pi_{\text{Strategy 2}} &= \sum_{i=1}^{2} [(P_i - C_i) \text{Min}(D_i, K_i)] - C_{\text{Strategy 2}} \sum_{i=1}^{2} K_i \\ \text{Take, } M_i &= \text{Min}(D_i, K_i) = \text{Production quantity of product i} \\ \text{Hence the simulated data based optimization model becomes.} \\ \text{Maximize} & \frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{2} [(P_i - C_i) M_{ij}] - C_{\text{Strategy 2}} \sum_{i=1}^{2} K_i \\ \text{Subject to:} & M_{ii} \leq d_{ii} \quad \forall i, j \end{split}$$

$$\begin{split} M_{ij} &\leq d_{ij} & \forall \ i,j \\ M_{ij} - K_i &\leq 0 & \forall \ i,j \\ M_{ij}, K_i &\geq 0 & \forall \ i,j \end{split}$$

Two Product, Flexible Plant:

This strategy has been discussed already in the previous sub-section.

3.3. Comparison between Analytical and Simulated Data Based Procedure:

To find optimal capacity levels and maximum profit and corresponding optimal capacity values for the three strategies discussed above, working has been done on two different parameter sets. The values have been generated by both analytical (wherever possible) and simulated data based procedure. In both the examples demands have been considered to be followed independent normal distribution with given parameters. Using these parameters 10,000 demand scenarios has been generated. Percent deviation has been calculated using the following formula:

Percent deviation = (Simulated data based result-Analytical result)×100/(Analytical result)

Example 1: Consider marginal cost of Capacity for any case = 4.

Data	Price	Cost	Salvage Value	Mean Demand	Std. dev. of Demand
Product 1	15	9	5	100	25
Product 2	13	8	3	200	40

Other Parameters are shown below:

Analytical	Case 1		Cas	Case 2		Case 3	
based results	Capacity	Profit	Capacity	Profit	Capacity	Profit	
Product 1	78.96	130.0	89.23	145.5	260.3		
Product 2	148.74	129.8	166.34	144.0			

Simulated data	Case 1		Case 2		Case 3	
based results	Capacity	Profit	Capacity	Profit	Capacity	Profit
Product 1	79.24	130.7	89.59	146.2	260.74	334.2
Product 2	148.25	129.4	166.03	143.6		

Percent	Case 1		Cas	se 2	Case 3	
deviation	Capacity	Profit	Capacity	Profit	Capacity	Profit
Product 1	0.35	0.54	0.40	0.48	0.17	
Product 2	-0.33	-0.31	-0.19	-0.28		

Example 2: Consider marginal cost of Capacity for any case = 4.

Other Parameters are shown below:

Data	Price	Cost	Salvage Value	Mean Demand	Std. dev. of Demand
Product 1	15	9	5	200	40
Product 2	13	8	3	100	25

Analytical	Case 1		Case 2		Case 3	
based results	Capacity	Profit	Capacity	Profit	Capacity	Profit
Product 1	166.34	288	182.77	312.7	260.3	
Product 2	67.96	56.2	78.96	65.0		

Simulated data	Case 1		Case 2		Case 3	
based results	Capacity	Profit	Capacity	Profit	Capacity	Profit
Product 1	166.05	288.2	182.79	312.6	260.32	434.6
Product 2	68.06	56.5	78.9	65.2		

Percent	Case 1		Cas	se 2	Case 3	
deviation	Capacity	Profit	Capacity	Profit	Capacity	Profit
Product 1	-0.17	0.07	0.01	-0.03	0.01	
Product 2	0.15	0.53	-0.08	0.31		

From the above examples one can conclude that the results found using simulated data based optimization procedures are very close to the results found using analytical procedures (deviations are less than 0.5% for most of the cases). Now, in the next sections, multivariate analysis and correlation will be introduced, and this becomes extremely difficult if not impossible to solve by analytical method and obtain closed form solution for optimum profit and capacity levels. Hence for the rest of the paper, whenever it has been required to maximize profit for the optimal capacity levels, simulated data based optimization has been used.

4. Multi Product Cases:

4.1. Methodology:

In this section multivariate normal demand distribution has been considered, so that it can capture the effects of correlation on profit level. Normal numbers have been generated by using variance-covariance matrix. The demands of the products $D_i \in R_+$ are random draws from a multivariate normal distribution function. For product i, realization of demand is d_i , the mean of the marginal distribution is μ_i , the variance is σ_i^2 , and the covariance of the joint distribution is $\sigma_{ik} = \rho_{ik}\sigma_i\sigma_k$, where $l \ge \rho_{ik} \ge -1$ for $i \ne k$.

For three products following correlated multivariate distribution, finite discretization of random parameter allows writing the expectation in the form of summation and makes the problem tractable. The random multivariate normal numbers have been produced by pre-multiplying a vector of random univariate normal numbers by the Cholesky decomposition of the Variance–Covariance matrix (\mathbf{V}) according to the formula:

$$Z = \mu + LX$$
 (8)

Where,

 \mathbf{Z} = a vector of random multivariate normal numbers

 μ = a vector of mean of the marginal distribution

 $\mathbf{X} = a$ vector of random univariate normal numbers

 \mathbf{L} = the Cholesky decomposition of the covariance matrix.

Here the values derived from the Cholesky decomposition have been stored in the lower triangle and main diagonal of a square matrix; elements in the upper triangle of the matrix are 0.

If variance–covariance matrix is real, symmetric and positive definite, then Cholesky decomposition exists.

Positive-definiteness:

An arbitrary matrix is positive definite if and only if all the principal sub-matrices have a positive determinant.

4.1.1. Cholesky decomposition Algorithm:

$V = LL^T$

Start with L=0

for i=1... *m do*

Subtract from $v_{i,i}$, the dot product of the *i*th row of *L* with *i*tself and set $l_{i,i}$ to be the square root of this.

for j=i+1,...,m

Subtract from $l_{i,j}$, the dot product of the *i*th and *j*th rows of L and set $l_{j,i}$ to be this result divided by $l_{i,i}$.

4.1.2. Example of Cholesky Decomposition:

Consider **V**, a $[3 \times 3]$ matrix, as given below:

$$\mathbf{V} = \begin{bmatrix} 10 & 2 & -4 \\ 2 & 15 & 1 \\ -4 & 1 & 6 \end{bmatrix}$$

Matrix is real, symmetric and positive definite. Hence Cholesky decomposition exists. The steps are shown below:

$$\mathbf{L^{1}}{=}\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix};$$

For i = 1, $v_{1,1} = 10$ and 1^{st} row of $\mathbf{L} = \begin{bmatrix} 0 & 0 & 0 \end{bmatrix}$; the dot product = $0 \times 0 + 0 \times 0 + 0 \times 0 = 0$; hence, $l_{1,1} = \sqrt{(v_{1,1} - dot \text{ product})} = \sqrt{(10 - 0)} = 3.16$.

$$\mathbf{L}^2 = \begin{bmatrix} 3.16 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix};$$

Given i = 1, for j = 2, 1st row of $\mathbf{L} = \begin{bmatrix} 3.16 & 0 & 0 \end{bmatrix}$ and 2nd row of $\mathbf{L} = \begin{bmatrix} 0 & 0 & 0 \end{bmatrix}$; the dot product = 3.16×0 + 0×0 = 0; hence $l_{2,1} = (v_{2,1} - \text{dot product})/l_{1,1} = 2/3.16 = 0.63$.

Given i = 1, for j = 3, 1st row of $\mathbf{L} = \begin{bmatrix} 3.16 & 0 & 0 \end{bmatrix}$ and 3rd row of $\mathbf{L} = \begin{bmatrix} 0 & 0 & 0 \end{bmatrix}$; the dot product = 3.16×0 + 0×0 = 0; hence $l_{3,1} = (v_{3,1} - \text{dot product})/l_{1,1} = -4/3.16 = -1.26$.

$$\mathbf{L}^{3} = \begin{bmatrix} 3.16 & 0 & 0 \\ 0.63 & 0 & 0 \\ -1.26 & 0 & 0 \end{bmatrix};$$

For i = 2, $v_{2,2} = 15$ and 2^{nd} row of $\mathbf{L} = \begin{bmatrix} 0.63 & 0 & 0 \end{bmatrix}$; the dot product = $0.63 \times 0.63 + 0 \times 0 + 0 \times 0 = 0.3969$; hence, $l_{2,2} = \sqrt{(v_{2,2} - \text{dot product})} = \sqrt{(15 - 0.3969)} = 3.82$.

$$\mathbf{L}^{4} = \begin{bmatrix} 3.16 & 0 & 0 \\ 0.63 & 3.82 & 0 \\ -1.26 & 0 & 0 \end{bmatrix}$$

Given i = 2, for j = 3, 2^{nd} row of **L** = [0.63 3.82 0] and 3^{rd} row of **L** = [-1.26 0 0]; the dot product = $0.63 \times (-1.26) + 3.82 \times 0 + 0 \times 0 = 0$; hence $l_{3,2} = (v_{3,2} - \text{dot product})/l_{2,2} = (1 - (-0.79))/3.82 = 0.47$.

$$\mathbf{L}^{5} = \begin{bmatrix} 3.16 & 0 & 0\\ 0.63 & 3.82 & 0\\ -1.26 & 0.47 & 0 \end{bmatrix};$$

For i = 3, $v_{3,3} = 6$ and 3^{rd} row of $\mathbf{L} = \begin{bmatrix} -1.26 & 0.47 & 0 \end{bmatrix}$; the dot product = $(-1.26) \times (-1.26) + 0.47 \times 0.47 + 0 \times 0 = 1.8$; hence, $l_{3,3} = \sqrt{(v_{3,3} - dot product)} = \sqrt{(6 - 1.8)} = 2.04$.

Finally $\mathbf{L} = \begin{bmatrix} 3.16 & 0 & 0\\ 0.63 & 3.82 & 0\\ -1.26 & 0.47 & 2.04 \end{bmatrix}$

Now, as per eq. (1) one only needs to generate \mathbf{X} , column vector of standard normal random numbers. Then by the use of eq. (1), each set of \mathbf{X} has been used generates one set of random numbers from multivariate normal distribution. In this way, 10,000 sets of samples have generated for the purpose. It has been seen that, with this large number the samples, statistics follow original distribution parameters.

Alike previous section, here also 10,000 demand data sets have been used for the optimization procedure. Models for simulated data based optimization are same as two product cases (section 3.3), except total number of products in these cases are 3. Hence, in this sub-section opportunity loss models for the above-mentioned strategies have been introduced.

4.2. Opportunity Loss Models:

Multi Product, Dedicated Plants, No Production Postponement:

 $\Pi_{\text{Strategy 1}} = \sum_{i=1}^{m} [(P_i - C_i) \text{Max}(D_i - K_i, 0) + (S_i - C_i) \text{Max}(K_i - D_i, 0)] + C_{\text{Strategy 1}} \sum_{i=1}^{m} K_i$ Hence the model becomes,

$$\begin{split} \text{Minimize} \quad & \frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{m} \left[(P_i - C_i) U_{ij} - (S_i - C_i) V_{ij} \right] + C_{\text{Strategy 1}} \sum_{i=1}^{m} K_i \\ \text{Subject to:} \quad & D_{ij} - K_i = U_{ij} - V_{ij} \quad \forall \ i, j \\ & U_{ij}, V_{ij}, K_i \geq 0 \qquad \forall \ i, j \end{split}$$

Multi Product, Dedicated Plants, Production Postponement:

 $\Pi_{\text{Strategy 2}} = \sum_{i=1}^{m} [(P_i - C_i) \text{Max}(D_i - K_i, 0)] + C_{\text{Strategy 2}} \sum_{i=1}^{m} K_i$ <u>Hence the model becomes.</u> Minimize $\frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{m} [(P_i - C_i) U_{ij}] + C_{\text{Strategy 2}} \sum_{i=1}^{m} K_i$

Subject to: $D_{ij} - K_i = U_{ij} - V_{ij} \quad \forall i, j$ $U_{ij}, V_{ij}, K_i \ge 0 \qquad \forall i, j$

Multi Product, Flexible Plant:

$$\begin{split} \Pi_{\text{Strategy }3} &= \sum_{i=1}^{m} [(P_i - C_i) \text{Max}(D_i - K_i, 0)] + C_{\text{Strategy }3} \text{K} \\ \underline{\text{Hence the model becomes,}} \\ \text{Minimize } & \frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{m} [(P_i - C_i) U_{ij}] + C_{\text{Strategy }3} \text{K} \\ \text{Subject to: } & \sum_{\substack{i=1 \\ U_{ij}}}^{m} D_{ij} - \text{K} = \sum_{i=1}^{m} (U_{ij} - V_{ij}) \quad \forall j \\ U_{ij}, V_{ij}, \text{K}_i \geq 0 \qquad \forall i, j \end{split}$$

As intuitive, capacity values in case of profit models and opportunity loss models are same.

4.3. Findings:

Variance and coefficient of variation represent two common measures of individual level of demand uncertainty. On the other hand change in correlation changes aggregate level demand uncertainty keeping variance unchanged. In the numerical analysis, considering three-product environment, the effect of these uncertainties on optimal expected total profit and corresponding total capacity level has been tried to capture for all the strategies discussed above. $\mu_i = 500$, $P_i = 80$, $C_i = 20$, $S_i = 5$ for all products and $C_{\text{Strategy S}} = 10$ for all strategies. However, incorporating differences in cost of capacities of dedicated and flexible plants can be easily done. To check the

effects of uncertainty following parameter sets have been considered: changes in coefficient of variation = $\{0.05, 0.1, 0.15, 0.2\}$, changes in variance = $\{2000, 4000, 6000, 8000, 10000\}$, changes in correlation = $\{0.99, 0.5, 0.25, 0, -0.25, -0.5\}$ for all products. The results are tabulated in Appendix A.1. The graphs have been shown in Appendix E. The observations have been discussed below:

- 1) Except when correlation is negative, the capacity of the flexible plant remains in between total capacity of the dedicated plants having no postponement and the same having production postponement, with last one giving the highest capacity. However, for highly positively correlated (0.99) demands, under postponement, the capacity of flexible plant becomes equal to aggregate capacity of dedicated plants. For negatively correlated demands, flexible plant optimal capacity is always the least. Intuitively, production postponement tends to increase the capacity as a result of elimination of overproduction, while the flexibility reduces the capacity due to pooling effect. For highly negatively correlated demands, aggregate demand variance almost reduces to zero and capacity approaches total mean demand value. For example in our case minimum possible correlation is 0.5, as variance–covariance matrix does not remain positive definite below this value. At this level of correlation, irrespective of variance, total demand realization becomes 1500. Hence there is no aggregate level of uncertainty at this value and flexible plant capacity also remains at 1500.
- 2) In terms of profit, flexible plant always remains the best choice, followed by dedicated plant with production postponement; dedicated plants having no postponement give the least profit. For negatively correlated demand benefit from flexible plant is intuitive as below average realized demand for one product has been compensated by the higher than average realized demand for another product. For example alike capacity, with correlation of -0.5, profit remains unaffected by the variance level. However, for highly positively correlated (0.99) demands, under postponement, the profit from flexible plant becomes equal to that of dedicated plant.
- For dedicated plant strategies, profit and capacity remains unaffected with the change in correlation coefficient. However, with the reduction in correlation flexible plant optimal profit increases and capacity decreases.

Now the effects of uncertainties have been discussed on the strategies for a) change in demand differential, b) change in price differential, c) change in price and d) change in capacity cost.

- a) When the effects of change in demand differential have been examined, mean demands for the products have been considered as follows: {500, 500, 500}; {400, 500, 600} and {250, 500, 750}. This helps to observe the effects on three levels of demand differential, {0, 200, 500}, average mean demand unchanged, where first one corresponds to the base case. Other values remain same: $P_i = 80$, $C_i = 20$, $S_i = 5$ and $C_{Strategy S} = 10$. The results have been tabulated in Appendix A.1. The graphs have been shown in Appendix E. it has been observed that, changes in demand differential, has no effect on optimal profit and capacity for any of the three strategies, but individual profits and capacities change.
- b) For examining the effects of change in price differential, prices for the products have been considered as follows: {80, 80, 80}; {60, 80, 100} and {40, 80, 120}. This helps to observe the effects on three levels of price differential, {0, 40, 80}, where first one corresponds to the base case and average product price remains unchanged. Other values remain same: $\mu_i = 500$, $C_i = 20$, $S_i = 5$ and $C_{\text{Strategy S}} = 10$. The results have been tabulated in Appendix A.2. The graphs have been shown in Appendix E. Increase in price differential decreases capacity for all three strategies. However, the effect of price differential on optimal total profit is not much.
- c) For observing the effects of change in price, three price levels, {40, 55, 80} have been considered, where all the products have same price. Other values remain same: $\mu_i = 500$, $C_i = 20$, $S_i = 5$ and $C_{\text{Strategy S}} = 10$. The results have been tabulated in Appendix A.3. The graphs have been shown in Appendix E. The effects of no postponement and postponement on optimal total capacity and profit have been discussed below.
 - 1) In case of dedicated plant with no production postponement strategy when price of the product is low (40), capacity is less than the expected total demand and capacity decreases with the increase in variance. Similarly, when price of the product is high (80), capacity is greater than the expected total demand and capacity increases with the increase in variance. The reason is quite simple; when price is low, then cost of overstocking (20 + 10 5 = 25) exceeds the cost of understocking (40 20 10 = 10) and the capacity is maintained at a lower side. With the increase in variance, expected loss from overstocking increases more compared to expected loss from understocking.

Hence the capacity also reduces. When the price is high exactly opposite happens (cost of overstocking = $25 < \cos t$ of understocking = 50). When cost of overstocking and cost of understocking are almost same (at price 55 the value is 25), capacity has been maintained near to expected demand value and capacity remains indifferent with the change in variance. However optimal expected profit always reduces with the increase in variance or coefficient of variation as with the increase in individual uncertainty both the expected understocking and expected overstocking cost increases. These results are in line with the analytical findings in two product case.

- 2) In case of production postponement, capacity increases and profit reduces with the increase in variance for both dedicated and flexible plants. However, when the cost of overcapacity and the cost of undercapacity both remain same, optimal total capacity level remains almost equal to total mean demand in both dedicated and flexible plants having production postponement and remains unaffected by the changes in variance and correlation. For example, when price is 55, cost of undercapacity = 55 - 20 - 10 = 25; when price is 40, cost of undercapacity = 40 - 20 - 10 = 10, where the cost of overcapacity = capacity cost = 10. In the first case total capacity is higher than mean demand, 1500. However, in the second case total capacity approaches the mean demand and remains unaffected by variance. The reason is quite simple. In case of production postponement there is no chance of overproduction, but there is always cost of underproduction due to capacity constraint and is same as undercapacity cost. As long as the cost of overcapacity does not exceed the cost of undercapacity, firm always gains from higher realized demand by maintaining higher capacity. At the same time, there is no loss from low realized demand except having idle capacity. But if the overcapacity cost is higher, the firm only tries to maintain capacity at mean demand level.
- d) For looking into the effects of change in capacity cost, three levels, {5, 10, 15} have been considered, where all the strategies have same capacity costs. Other values remain same: $\mu_i = 500$, $P_i = 80$, $C_i = 20$ and $S_i = 5$. The results have been tabulated in Appendix A.4. The graphs have been shown in Appendix E. Increase in cost of capacity reduces both capacity and profit in all cases.

In all the cases discussed above, change in correlation has no effect on unmet demand percentage for dedicated as well as flexible plant. However, with highly negatively correlated demand, when aggregate demand variance becomes negligible, flexible plant has no unmet demand (See Appendix C).

As flexible plant is effective only if there is production postponement, an index called 'PdPPF Index' has been introduced to check the effect of production postponement on flexible plant profit where 'PdPPF' stands for 'Production Postponement effect on Flexible plant'. The index is calculated as below:

$PdPPF Index = \frac{Profit_{Dedicated Plant, Postponement} - Profit_{Dedicated Plant, No Postponement}}{Profit_{Flexible Plant} - Profit_{Dedicated Plant, No Postponement}} \times 100\%$

A reduction in PdPPF index with the increase in a particular parameter indicates that the abovementioned effect reduces and suggests that manufacturer can invest more in product flexible technology. Hence, this PdPPF index can also be considered as the proxy of the value of product flexibility. Although, profit decreases with the increase in variance for all strategies, PdPPF index remains unaffected in variance or coefficient of variation; which means, the value of product flexibility has not been affected by the individual level demand uncertainty. However, PdPPF index decreases with the decrease in correlation. With the increase in negative correlation, manufacturer's incentive to invest in product flexibility increases. PdPPF index also decreases with the value of product flexibility increases in marginal cost of capacity. So it can be concluded that the value of product flexibility increases in price differential. Product flexibility becomes more effective when higher price differential or lower marginal cost of capacity has been combined with lower correlation. With the increase in negative correlation, the price differential effect diminishes with increase in negative correlation, but the effect of marginal cost of capacity increases with the increases with the increase in negative correlation, but the effect of marginal cost of capacity increases with the increase in negative correlation. The explanations have been given below (For PdPPF Index values see Appendix B.5):

a) With the change in price differential, the total profits of dedicated plant (both postponement and no postponement) strategies have not been affected much. They also remain unaffected with the change in correlation. However, the profit of flexible plant increases with decrease in correlation due to pooling effect. With the increase in price differential, this additional gain from flexibility becomes less effective. In other words, correlation effect acts better in flexible plant when price differential is low. The explanation is simple. With the increase in price differential, the flexible firm increases the option to allocate more of it's resource to the high price product, so that it can always

meet the demand of high price product, even at the cost of low price product. As a result in case of high correlation also, flexible firm profit is more compared to dedicated plant with postponement. As a result, for different price differentials, PdPPF Index converges with decrease in correlation. The same can be observed in graph also (See Appendix E, Graph 6(a)).

- b) With the reduction in correlation, PdPPF Index decreases irrespective of the price of the product. However, with very high correlation (0.99) they converge at value =1. The reason is quite intuitive. At 0.99 correlation value, there is no additional gain from flexible plant over dedicated plant with postponement. So the optimal profit in both cases remains same and the PdPPF Index value becomes unity. With the change in correlation the optimal profit in dedicated strategies do not change, but the profit of flexible plant increases due to pooling effect. Hence, PdPPF Index decreases. However, with the reduction in product price, this additional gain from flexibility becomes less effective (with low price product the flexible plant profit range reduces). In other words, correlation effect acts better in flexible plant when prices of the products are high. But, at the same time, with the decrease in price the extra benefit of postponement reduces as the cost of understocking reduces and capacity has been maintained at lower side. Hence, the difference between profits in dedicated plant strategies reduces. As a result with the decrease in correlation PdPPF Index diverges. The same can be observed in graph also (See Appendix E, Graph 6(b)).
- c) In case of change in capacity cost, the structure of the graph and explanation on PdPPF Index is same as the effect of change in price. The same can be observed in graph also (See Appendix E, Graph 6(c)).

5. Service level constraint in multi product case:

In today's customer oriented business, maximizing profit is not the only target for firms. They also need to consider the service level as a satisfying objective. Here, service level means the expected number of cases where the demand has been met. Maximizing expected profit being the main objective of the firm, service level objective has been taken as constraint to the models. The aggregate service level has been calculated by averaging individual service levels, based on the assumption that same cost of stock-out occasions for products.

5.1. Mixed Integer Models for Satisfying Aggregate Service Level:

When constraints have been added for satisfying service level in models presented in section 3.2.2 one can go for the following argument. Given A = a large number, I = binary variable, D = demand of product, Z = Sales level, if $A*I \ge D - Z$, then

$$D - Z > 0 \rightarrow I = 1$$
$$D - Z = 0 \rightarrow I = 0, 1$$
$$D - Z < 0 \rightarrow I = 0, 1$$

Then, if "total I \leq a given value" has been used as a constraint, it will try to assign zero to I values, whenever required. In other words, it will try to make D – Z \leq 0. So unmet demand instances will be reduced upto the required level.

Multi Product, Dedicated Plants, No Production Postponement (Service level \geq 70%):

Maximize	$\frac{1}{n}\sum_{j=1}^{n}\sum_{i=1}^{m}\left[(P_{i}-S_{i})Z_{ij}\right]-\sum$	$C_{i=1}^{m}(C_{\text{Strategy }1} + C_i - S_i)K_i$
Subject to:	$\begin{array}{l} Z_{ij} \leq d_{ij} \\ Z_{ij} - K_i \leq 0 \\ d_{ij} - Z_{ij} - A \times I_{ij} \leq 0 \\ \sum_{j=1}^n \sum_{i=1}^m I_{ij} \leq 0.3mn \end{array}$	∀ i, j ∀ i, j ∀ i, j
	$I_{ij} \text{ binary} Z_{ij}, K_i \ge 0 A = Big positive number$	∀i,j ∀i,j r

Multi Product, Dedicated Plants, Production Postponement (Service level \geq 90%):

Maximize	$\frac{1}{n} \sum_{j=1}^{n} \sum_{i=1}^{m} \left[(P_i - C_i) M_{ij} \right] - 0$	$\sum_{\text{Strategy } 2} \sum_{i=1}^{m} K_i$
Subject to:	$M_{ij} \leq d_{ij}$	∀ i, j
	$M_{ij} - K_i \leq 0$	∀ i, j
	$d_{ij} - M_{ij} - A \times I_{ij} \leq 0$	∀ i, j
	$\sum_{j=1}^n \sum_{i=1}^m I_{ij} \leq 0.1 mn$	
	I _{ij} binary	∀ i, j
	M_{ij} , $K_i \ge 0$	∀ i, j
	A = Big positive number	r

Multi Product, Flexible Plant, Production Postponement (Service level \geq 90%):

Variable	$M_{ij} \geq 0$	∀ i, j;	$K \ge 0;$	I _{ij} binary ∀i,j
Maximize	$\frac{1}{n}\sum_{j=1}^{n}\sum_{i=1}^{m}\left[\left(1\right)$	$P_i - C_i M_i$	$_{ij}] - C_{Strategy}$	₃ K

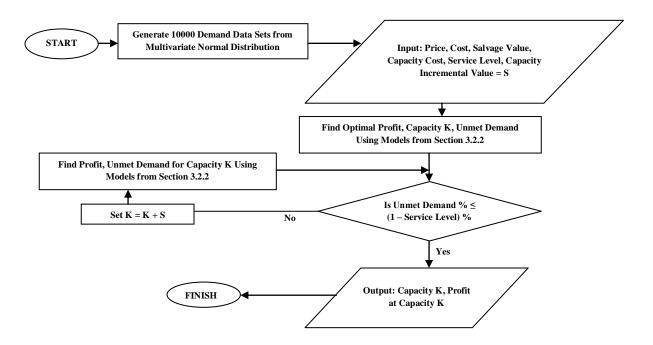
Subject to:	$M_{ij} \leq d_{ij}$	∀i,j
	$\sum_{i=1}^{m} M_{ij} - K_i \leq 0$	∀i,j
	$d_{ij} - M_{ij} - A \times I_{ij} \le 0$	∀ i, j
	$\sum_{j=1}^n \sum_{i=1}^m I_{ij} \leq 0.1 mn$	
	I _{ij} binary	∀i,j
	$M_{ij}, K_i \geq 0$	∀i,j
	A = Big positive number	

When one goes for satisfying individual service level only fourth constraint changes as below:

$\sum_{j=1}^n I_{ij} \leq 0.3n$	\forall i (When required service level \geq 70%) and,
$\sum_{j=1}^n I_{ij} \leq 0.1n$	\forall i (When required service level \geq 90%)

The problem with MIP models discussed above is that in many cases it can not perform with more than 100 demand data sets. So alternate approach has been adopted, which has been discussed next.

5.2. Simulation Diagram for Checking Required Service Level:



5.3. Results and Findings:

To compare the results with section 4.3 same parameter values have been kept. The observations have been discussed below (For PdPPF Index values with SLC see Appendix B.5):

1) Observations regarding PdPPF Index: PdPPF decreases with decrease in correlation in both SLC and without SLC. At same correlation level, with SLC, PdPPF increases with decrease in price differential, increase in price and decrease in capacity cost. However, without SLC PdPPF values converges at highly negative correlation when change in price differential happens, and it converges at highly positive correlation when change in price or change in capacity cost happens. On the other hand, when service level constraint has been imposed, PdPPF Index diverges with decrease in correlation for different price differential levels as well as for different price levels. However, with different capacity cost PdPPF Index remains parallel. The explanations have been given below:

- a) As the service level constraints in any of the price differential levels are not violated much, the differences between profits in dedicated plant strategies are alike in this case compared to the cases without SLC. Hence, when operated under SLC, for different price differential levels, PdPPF Index converges with decrease in correlation. The same can be observed in graph also (See Appendix E, Graph 7(a)).
- b) With the decrease in the product price, rate of decrease in Unmet Demand Percentage (hence, UD %) is more in case of dedicated plant with postponement compared to flexible plant. On the other hand, dedicated plant with no postponement has been affected most with service level constraint. As a result, the less the product price, the more profit reduction happens for dedicated plant strategies to fulfill the service level requirement. When price values are 80 and 55, flexible plant service levels are not violated and, in case of price = 40, a small reduction in profit happens to maintain the service level. Hence, even with correlation = 0.99, the profits between dedicated plant with postponement and flexible plant differs and this difference increases with decrease in product price. However, with the decrease in correlation, dedicated plant (both postponement and no postponement) profits with SLC remain same as optimal profit (without SLC) and UD % do not change in correlation for dedicated plants. As with the reduction in product price, additional gain from flexibility becomes less effective, the denominator of the PdPPF Index shows less increase with the reduction in correlation when product price is low. On the other hand, the reduction in differences between profits in dedicated plant strategies are less compared to the cases without SLC. As a result, as product price decreases, the PdPPF Index decreases less with the reduction in correlation. In other words, when

operated under SLC, for different product prices, PdPPF Index diverges with increase in correlation. The same can be observed in graph also (See Appendix E, Graph 7(b)).

c) In case of change in capacity cost, the reason for non convergence at correlation = 0.99 is same as the effect of price change. However, as the service level constraints in any of the capacity cost levels are not violated much, the reduction in differences between profits in dedicated plant strategies are alike in this case compared to the cases without SLC. Hence, when operated under SLC, for different capacity cost levels, PdPPF Index remains parallel with the change in correlation. The same can be observed in graph also (See Appendix E, Graph 7(c)).

2) Observations regarding reduction in profit from imposing Service Level Constraint (SLC):

Reduction in Profit Percentage = RP % = [(Profit with SLC – Profit without SLC) ×100/ Profit without SLC]. More negative value of reduction in profit means more decrease in profit with SLC (See Appendix D).

- a) In case of dedicated plant with postponement, correlation has no effect on RP %. The RP % decreases with correlation in case of flexible plant.
- b) With increase in variance and decrease in price of products, RP % decreases for both dedicated and flexible plant. However the effect is less in case of flexible plant.
- c) Change in price differential or change in capacity cost has little effect on the RP %.

6. Conclusion:

This paper deals with the optimal capacity planning under demand uncertainty. Simulated data based optimization procedure used in this paper helped to solve the multi-product two stage stochastic linear programming which is otherwise analytically intractable. The effect of production postponement increases profit, but flexible plant may generate higher profit compared to dedicated plants depending on the cost of flexible technology. For dedicated plant strategies, profit and capacity remains unaffected with the change in correlation coefficient. However, with the reduction in correlation flexible plant optimal profit increases and capacity decreases. On the other hand, change in demand differential or price differential has no effect on aggregate capacity or profit for any of the three strategies. But profit reduces with the reduction in price or increase in capacity cost.

The PdPPF index introduced in the paper is helpful in deciding on choice between dedicated and product flexible plant. The value of flexibility has not been affected by the change in individual demand uncertainty, but effectiveness of product flexibility increases with negatively correlated demands. The change in demand differential has no effect on aggregate capacity or aggregate profit level in any of the three strategies, however, increase in price differential or decrease in marginal cost of capacity increases the value of flexibility.

Change in correlation has no effect on unmet demand percentage for dedicated as well as flexible plant. However, with highly negatively correlated demand, when aggregate demand variance becomes negligible, flexible plant has no unmet demand. On the other hand, when service level constraint has been imposed, PdPPF Index diverges with decrease in correlation for different price differential levels as well as for different price levels. However, with different capacity cost PdPPF Index remains parallel.

The main contribution of this paper is threefold. First, the procedure of finding optimal profit and capacity has been developed for dedicated and flexible plant facing multivariate correlated demand distribution, which is otherwise analytically intractable. Second, PdPPF Index has been introduced as a proxy for value of product flexibility to find several meaningful insights based on the changes in various parameters. Third, the service level objective has been added to look into the problem from multi-objective angle.

Several extensions to the models are possible. We are currently working on price dependent demand scenario to accommodate price postponement into our model and observe the effect of product substitutability. Another extension on which we are also working is to extend the models for multi-period scenario.

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Appendix A: Results

Appendix A.1:

Parameter values: Price = 80, Cost = 20, Salvage value = 5 for each of the three products; Marginal cost of capacity = 10 for any type of plant $\sigma/\mu = \{0.05, 0.1, 0.15, 0.2\}, \sigma^2 = \{2000, 4000, 6000, 8000, 10000\}$

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Mean Demand = {500, 500, 500}	Mean Demand = {400, 500, 600}	Mean Demand = {250, 500, 750}				
Change in σ/μ , $\rho = 0.99$	Change in σ/μ , $\rho = 0.99$	Change in σ/μ , $\rho = 0.99$				
1532 1574 1573 72965 73891 73895	1532 1572 1571 72891 73822 73825	1531 1573 1573 72888 73823 73826				
1562 1642 1641 70828 72661 72668	1566 1648 1648 71000 72845 72853	1565 1643 1642 70911 72751 72758				
1599 1719 1719 68973 71741 71752	1597 1715 1714 68872 71621 71633	1601 1721 1720 69028 71794 71804				
1623 1779 1777 66731 70355 70370	1631 1791 1792 66986 70656 70671	1628 1789 1787 33779 70464 70476				
	Change in σ^2 , $\rho = 0.99$					
	1556 1628 1627 71335 72970 72977	1556 1630 1630 71349 72996 73304				
	1579 1682 1682 69680 72019 72028					
	1604 1731 1731 68678 71578 71590					
		1614 1759 1759 67832 71083 71096				
	1633 1793 1790 66828 70553 70569					
	Change in σ/μ , $\rho = 0.5$					
		1532 1571 1559 72965 73882 74068				
1565 1645 1621 70929 72771 73206	1566 1646 1619 71019 72856 73262	1567 1644 1622 70942 72788 73167				
1597 1716 1677 68923 71662 72269	1598 1716 1676 69039 71764 72374	1597 1716 1680 68950 71696 72249				
1629 1787 1734 66855 70522 71370	1633 1796 1743 66907 70625 71426	1631 1789 1743 66818 70519 71250				
Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$				
	1558 1632 1608 71360 73026 73384					
1581 1685 1650 69785 72123 72648	1582 1683 1648 69934 72238 72764	1582 1684 1650 69872 72196 72720				
1598 1724 1681 68577 71428 72074	1596 1723 1681 68523 71377 72027	1600 1725 1684 68617 71485 72124				
	1616 1761 1710 67684 70984 71709					
1629 1787 1734 66855 70522 71370	1631 1792 1735 66880 70571 71394	1629 1789 1736 66939 70598 71418				
Change in σ/μ , $\rho = 0.25$	Change in σ/μ , $\rho = 0.25$	Change in σ/μ , $\rho = 0.25$				
	1531 1571 1550 72909 73826 74154					
	1562 1642 1600 70826 72660 73314					
	1593 1714 1650 68659 71429 72391					
1628 1793 1706 66824 70515 71858	1633 1793 1705 67033 70721 72038	1627 1792 1716 66763 70451 71643				
Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$				
	1558 1631 1592 71379 73022 73621					
	1582 1686 1630 69785 72131 72967					
		1598 1722 1655 68712 71541 72600				
		1615 1760 1685 67691 70989 72158				
		1625 1786 1707 66660 70351 71649				
Change in σ/μ , $\rho = 0$	Change in σ/μ , $\rho = 0$	Change in σ/μ , $\rho = 0$				
1533 1573 1542 72979 73906 74383	<u>1532</u> <u>1572</u> <u>1543</u> <u>72973</u> <u>73893</u> <u>74355</u>	1531 1573 1545 72953 73872 74294				
1566 1647 1587 70949 72804 73735	1563 1644 1583 70845 72691 73634	1562 1643 1586 70858 72700 73543				
		1598 1719 1639 68856 71628 72883				
		1631 1792 1685 66915 70595 72269				
Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$				
		1557 1629 1575 71340 72982 73816				
	1582 1684 1605 69845 72180 73388					
		1600 1724 1630 68645 71509 72972				
		1615 1762 1652 67707 71018 72734				
1626 1791 1666 66631 70340 72223	1626 1786 1663 66725 70395 72288	1627 1791 1665 66705 70407 72299				

	Change in σ/μ , $\rho = -0.25$					
1531 1573 1529 72933 73858 74526	1532 1573 1531 72974 73897 74547	1532 1572 1535 72937 73864 74439				
1565 1646 1560 70958 72799 74145	1563 1645 1560 70893 72744 74047	1565 1645 1569 70982 72813 73961				
		1596 1716 1602 68795 71577 73319				
1629 1788 1617 66910 70563 73195	1628 1788 1620 66751 70347 73032	1629 1788 1640 66845 70511 72779				
Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$				
1557 1631 1553 71330 72988 74198	1557 1630 1553 71268 72940 74147	1558 1629 1553 71349 72996 74205				
1581 1683 1574 69810 72143 73832	1580 1683 1573 69801 72140 73819	1582 1684 1574 69855 72188 73868				
1598 1723 1589 68578 71434 73487	1602 1724 1592 68757 71606 73663	1601 1727 1592 68697 71556 73628				
		1612 1757 1605 67680 70947 73305				
1629 1788 1617 66910 70563 73195	1632 1791 1618 66781 70494 73179	1628 1793 1619 66826 70504 73161				
Change in σ/μ , $\rho = -0.5$	Change in σ/μ , $\rho = -0.5$	Change in σ/μ , $\rho = -0.5$				
1532 1573 1500 72941 73867 75000	1532 1572 1508 72949 73876 74871	1531 1572 1520 72931 73850 74649				
1565 1644 1500 70917 72765 75000	1565 1643 1516 70913 72755 74735	1563 1644 1541 70881 72720 74312				
1598 1717 1500 68840 71619 75000	1596 1718 1525 68868 71636 74614	1595 1717 1562 68919 71666 74046				
1628 1786 1500 66868 70523 75000	1630 1788 1533 66811 70485 74497	1627 1791 1584 66870 70536 73728				
Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$				
1557 1629 1500 71339 72988 75000	1558 1630 1500 71335 72986 75000	1558 1630 1500 71324 72978 75000				
1581 1685 1500 69790 72135 75000	1582 1683 1500 69822 72163 75000	1580 1684 1500 69840 72158 75000				
1603 1724 1500 68645 71521 75000	1601 1725 1500 68688 71533 75000	1599 1722 1500 68711 71537 75000				
1615 1761 1500 67655 70961 75000	1614 1758 1500 67745 71015 75000	1616 1762 1500 67625 70957 75000				
1628 1786 1500 66868 70523 75000	1630 1790 1500 66795 70487 75000	1628 1790 1500 66839 70509 75000				

Appendix A.2:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5, σ^2 = 10000 for each of the three products; Marginal cost of capacity = 10 for any type of plant; ρ = {0.99, 0.5, 0.25, 0, -0.25, -0.5}

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Price = {80, 80, 80}							Price = {60, 80, 100}							Price = {40, 80, 120}				
	Change in p						Change in p						Change in p					
1623	1783	1782	66686	70334	70349	1614	1775	1699	66687	70314	70929	1563	1725	1500	67613	70869	72539	
1629	1787	1734	66855	70522	71370	1611	1772	1659	66897	70443	71727	1566	1726	1501	67807	71060	73160	
1628	1793	1706	66824	70515	71858	1617	1776	1640	66992	70580	72301	1566	1728	1502	67670	70938	73284	
1626	1791	1666	66631	70340	72223	1619	1779	1618	67088	70674	72894	1566	1723	1501	67700	70945	73619	
1629	1788	1617	66910	70563	73195	1616	1778	1580	66923	70511	73373	1565	1724	1500	67822	71077	74116	
1628	1786	1500	66868	70523	75000	1619	1782	1500	66965	70589	75031	1566	1727	1500	67596	70870	74970	

Appendix A.3:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5 for each of the three products; Marginal cost of capacity = 10 for any type of plant; $\sigma^2 = \{2000, 4000, 6000, 8000, 10000\}$

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Price = {80, 80, 80}							Price = {55, 55, 55}							Price = {40, 40, 40}					
	Cha	nge i	n σ², ρ	= 0.99			Cha	inge i	n σ², ρ	Change in σ^2 , $\rho = 0.99$									
1558	1631	1631	71387	73039	73045	1501	1576	1576	34802	35898	35903	1425	1501	1501	13401	13932	13935		
1581	1683	1682	69781	72106	72116	1500	1607	1607	33769	35288	35295	1391	1500	1500	12714	13465	13470		
1596	1726	1724	68551	71424	71437	1498	1630	1630	32894	34737	34747	1368	1496	1495	12226	13130	13136		
1619	1762	1761	67742	71054	71069	1500	1652	1652	32174	34325	34336	1349	1497	1498	11830	12867	12875		
1623	1783	1782	66686	70334	70349	1500	1672	1671	31474	33922	33934	1328	1499	1499	11431	12596	12603		

Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$					
	1502 1577 1563 34840 35930 36226	1424 1500 1500 13397 13925 14127					
1581 1685 1650 69785 72123 72648	1500 1608 1587 33718 35253 35668	1391 1497 1500 12733 13474 13753					
1598 1724 1681 68577 71428 72074	1502 1634 1610 32910 34783 35312	1366 1498 1498 12235 13139 13476					
	1502 1653 1626 32234 34381 34962						
1629 1787 1734 66855 70522 71370	1502 1672 1641 31587 33996 34653	1334 1502 1499 11475 12647 13089					
Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$						
	1500 1577 1554 34831 35915 36389						
		1392 1499 1499 12769 13501 13945					
	1501 1631 1593 32880 34754 35557						
	1497 1650 1606 32097 34265 35195						
	1500 1669 1622 31511 33929 34993						
$Change in \sigma^2, \rho = 0$	Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$					
	1500 1576 1544 34845 35927 36581						
	1499 1607 1562 33686 35223 36173						
	1498 1631 1575 32802 34681 35875						
	1498 1653 1586 32099 34276 35634						
	1503 1669 1601 31586 33993 35491						
	Change in σ^2 , $\rho = -0.25$						
	1501 1576 1531 34841 35921 36863						
	1498 1605 1543 33711 35233 36566						
	1499 1631 1552 32852 34724 36375						
	1499 1649 1560 32132 34285 36175						
	1503 1671 1569 31589 34000 36104						
Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$						
	1500 1576 1500 34814 35898 37500						
	1500 1607 1500 33724 35248 37500						
	1500 1628 1500 32890 34745 37500						
	1499 1648 1500 32194 34328 37500						
1628 1786 1500 66868 70523 75000	1502 1672 1500 31460 33910 37500	1330 1499 1500 11442 12608 15000					

<u>Appendix A.4:</u> Parameter values: Mean demand = 500, Price = 80, Cost = 20, Salvage value = 5, $\sigma^2 = 10000$ for each of the three products; $\rho = \{0.99, 0.5, 0.25, 0, -0.25, -0.5\}$

→ Horizontally { Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3 }

Capacity cost = 5							Capacity cost = 10							Capacity cost = 15					
	Change in ρ						Change in p						Change in p						
1681	1912	1911	74632	79291	79300	1623	1783	1782	66686	70334	70349	1579	1706	1706	58809	61816	61835		
1681	1912	1833	74808	79457	79951	1629	1787	1734	66855	70522	71370	1577	1704	1668	58916	61886	62920		
1690	1922	1795	75345	79997	80795	1628	1793	1706	66824	70515	71858	1575	1698	1642	58873	61819	63461		
1691	1917	1737	75171	79812	81000	1626	1791	1666	66631	70340	72223	1577	1701	1617	58839	61809	64244		
1687	1917	1661	75144	79784	81439	1629	1788	1617	66910	70563	73195	1574	1699	1579	58789	61747	65146		
1684	1907	1500	75203	79763	82500	1628	1786	1500	66868	70523	75000	1578	1701	1500	58783	61764	67500		

Appendix B: Results with Service Level Constraint

Appendix B.1:

Parameter values: Price = 80, Cost = 20, Salvage value = 5 for each of the three products; Marginal cost of capacity = 10 for any type of plant $\sigma^2 = \{2000, 4000, 6000, 8000, 10000\}$

Demand = {500, 500, 500}	Demand = {400, 500, 600}	Demand = {250, 500, 750}					
Change in σ^2 , $\rho = 0.99$	Change in σ^2 , $\rho = 0.99$	Change in σ^2 , $\rho = 0.99$					
1589 1692 1661 71277 72848 72987	1587 1686 1656 71248 72794 72927	1588 1690 1630 71182 72759 72937					
1608 1742 1742 69797 72020 72030	1610 1742 1742 69609 71893 71903	1612 1760 1680 69827 71980 72207					
1625 1811 1779 68396 71059 71201	1633 1818 1787 68702 71387 71530	1628 1814 1723 68620 71292 71526					
	1648 1851 1822 67517 70696 70821						
	1660 1904 1843 66950 70293 70533	1659 1912 1792 66718 70146 70472					
Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$					
	1589 1690 1638 71191 72781 73260						
	1612 1752 1680 69746 71967 72623						
	1629 1811 1711 68548 71222 72030						
	1650 1852 1744 67814 70949 71858						
1652 1907 1791 66457 69868 70939	1661 1902 1770 66827 70273 71337	1663 1913 1738 67023 70408 71538					
Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$					
	1590 1691 1622 71325 72887 73583						
	1610 1773 1661 69725 71843 72888						
	1629 1815 1689 68483 71176 72396						
	1649 1852 1715 67669 70815 72168						
1654 1905 1761 66719 70096 71621	1667 1908 1734 66698 70136 71717	1656 1906 1704 66690 70095 71721					
Example 1 Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$					
	1589 1690 1606 71319 72877 73822						
	1612 1761 1634 69716 71885 73242						
	1629 1815 1659 68549 71235 72900						
	1647 1849 1682 67686 70820 72681						
	1658 1908 1697 66774 70191 72385						
Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$					
	1587 1690 1583 71196 72768 74025						
	1612 1765 1605 69700 71892 73746						
	1630 1816 1620 68599 71288 73528						
	1645 1851 1637 67556 70719 73255						
	1672 1912 1648 66942 70373 73294						
Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$					
	1588 1690 1500 71239 72809 75000						
	1611 1756 1500 69743 71967 75000						
	1630 1815 1500 68613 71283 75000						
	1646 1852 1500 67631 70772 75000						
1648 1854 1500 67590 70753 75000	1660 1911 1500 66745 70168 75000	1661 1914 1500 66725 70174 75000					

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Appendix B.2:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5, σ^2 = 10000 for each of the three products; Marginal cost of capacity = 10 for any type of plant; ρ = {0.99, 0.5, 0.25, 0, -0.25, -0.5}

→ Horizontally { Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3 }

	Price = {80, 80, 80}					Price = {60, 80, 100}					Price = {40, 80, 120}						
	Change in ρ , $\sigma^2 = 10000$						Change in ρ , $\sigma^2 = 10000$					Change in ρ , $\sigma^2 = 10000$					
1658	1912	1851	66958	70342	70583	1685	1898	1703	66357	69971	70921	1684	1893	1676	66698	70098	72076
1652	1907	1791	66457	69868	70939	1682	1914	1670	66737	70271	71943	1682	1894	1650	66789	70126	72569
1654	1905	1761	66719	70096	71621	1679	1898	1642	66602	70141	72210	1689	1890	1625	66851	70297	73059
1661	1909	1699	66849	70256	72408	1668	1906	1614	66759	70153	72734	1688	1899	1596	67132	70484	73677
1669	1899	1647	66791	70259	73122	1687	1899	1583	66722	70288	73486	1702	1896	1589	66727	70244	73771
1648	1854	1500	67590	70753	75000	1676	1908	1500	66721	70181	74975	1685	1886	1500	66940	70330	75006

Appendix B.3:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5 for each of the three products; Marginal cost of capacity = 10 for any type of plant; $\sigma^2 = \{2000, 4000, 6000, 8000, 10000\}$

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Price = {80, 80, 80}	Price = {55, 55, 55}	Price = {40, 40, 40}
Change in σ^2 , $\rho = 0.99$	Change in σ^2 , $\rho = 0.99$	Change in σ^2 , $\rho = 0.99$
1589 1692 1661 71277 7284872987	1590 1698 1579 34246 35385 35911	1575 1680 1590 12286 13091 13703
1608 1742 1742 69797 7202072030	1618 1757 1607 32939 34615 35204	1623 1772 1621 10898 12173 13200
1625 1811 1779 68396 7105971201	1642 1812 1630 32049 34067 34766	1638 1810 1651 10145 11700 12761
1646 1845 1813 67799 7088971013		
	1680 1899 1670 30568 33111 34010	1665 1888 1678 8942 10801 12194
Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$	Change in σ^2 , $\rho = 0.5$
	1592 1698 1564 34295 35428 36245	
	1619 1757 1588 32964 34630 35641	
	1640 1811 1608 31985 34015 35204	
	1650 1863 1629 31449 33581 34974	
1652 1907 1791 66457 6986870939		
Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$	Change in σ^2 , $\rho = 0.25$
1587 1690 1621 71211 7277773483	1591 1698 1555 34262 35411 36405	
1612 1754 1660 69708 7194072898		
1632 1817 1688 68644 7132572538	1648 1809 1591 31911 34019 35515	1641 1810 1591 10128 11705 13491
1648 1851 1714 67783 7091372269	1652 1861 1610 31357 33531 35251	1646 1857 1616 9533 11141 13153
1654 1905 1761 66719 7009671621	1683 1904 1626 30553 33143 35052	1671 1901 1619 8886 10759 13064
Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$	Change in σ^2 , $\rho = 0$
1587 1691 1606 71278 7283973785	1590 1695 1544 34230 35376 36565	1575 1680 1560 12286 13090 14205
1610 1755 1637 69692 7191773263	1622 1759 1565 33009 34696 36235	1601 1749 1558 11196 12320 13987
1631 1815 1661 68584 7126972910		
1645 1846 1678 67615 7073872599	1651 1865 1588 31378 33526 35676	1648 1859 1586 9563 11156 13554
	1683 1911 1600 30528 33060 35495	1671 1901 1618 8881 10750 13315
Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$	Change in σ^2 , $\rho = -0.25$
1588 1690 1584 71271 7283274097	1590 1697 1531 34254 35399 36868	1575 1678 1530 12273 13095 14501
1611 1755 1604 69760 7198073782	1621 1757 1544 32984 34672 36598	1603 1770 1559 11199 12166 14201
1629 1814 1622 68626 7129873528	1650 1811 1555 31910 34041 36353	1638 1809 1558 10170 11713 14097
	1649 1863 1562 31393 33531 36238	
	1680 1911 1572 30457 32974 36034	
Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$	Change in σ^2 , $\rho = -0.5$
1588 1691 1500 71235 7280975000	1588 1695 1500 34270 35395 37500	1575 1680 1500 12294 13096 15000
	1620 1760 1500 32968 34646 37500	
	1650 1812 1500 31938 34052 37500	
1646 1859 1500 67620 70724 75000	1649 1861 1500 31329 33500 37500	1646 1858 1500 9593 11187 15000
1648 1854 1500 67590 70753 75000	1680 1910 1500 30437 32974 37500	1680 1889 1500 8757 10844 15000

Appendix B.4:

Parameter values: Mean demand = 500, Price = 80, Cost = 20, Salvage value = 5, $\sigma^2 = 10000$ for each of the three products; $\rho = \{0.99, 0.5, 0.25, 0, -0.25, -0.5\}$

	Capacity Cost = 5						Capacity Cost = 10					Capacity Cost = 15					
	Change in ρ , $\sigma^2 = 10000$						Change in ρ , $\sigma^2 = 10000$					Change in ρ , $\sigma^2 = 10000$					
1686	1686 1916 1912 74899 79572 79580					1658	1912	1851	66958	70342	70583	1671	1911	1700	58724	60900	62070
1692	1919	1847	75387	80021	80537	1652	1907	1791	66457	69868	70939	1662	1890	1664	58198	60586	62618
1686	1915	1792	74987	79636	80449	1654	1905	1761	66719	70096	71621	1661	1898	1640	58276	60552	63292
1691	1919	1743	75288	79936	81091	1661	1909	1699	66849	70256	72408	1665	1914	1621	58372	60571	64140
1686	1918	1670	75119	79761	81408	1669	1899	1647	66791	70259	73122	1666	1903	1584	58439	60735	65197
1686	1916	1500	75099	79726	82500	1648	1854	1500	67590	70753	75000	1667	1910	1500	58453	60648	67500

→ Horizontally {Capacity 1, Capacity 2, Capacity 3, Profit 1, Profit 2, Profit 3}

Appendix B.5:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5 for each of the three products; Marginal cost of capacity = 10 for any type of plant; $\sigma^2 = \{2000, 4000, 6000, 8000, 10000\}$

→ Horizontally $\rho = \{0.99, 0.5, 0.25, 0, -0.25, -0.5\}$

	PdPPF Index without Service Level Constraint																
	Pri	ce = {8	80, 80,	80}		Price = {55, 55, 55}					Price = {40, 40, 40}						
	(Chang	ge in σ	2		Change in σ^2					Change in σ ²						
99.64 8	81.91	73.57	65.69	57.81	45.04	99.55 78.64 69.58 62.33 53.41 40.36 9					99.44	72.33	59.51	53.59	45.42	32.85	
99.57 8	81.66	73.53	66.39	58.01	45.01	99.54	78.72	70.18	61.80	53.31	40.36	99.34	72.65	62.24	53.54	45.19	32.80
99.55 8	81.53	73.79	65.67	58.18	45.26	99.46	77.98	70.00	61.15	53.14	40.24	99.34	72.84	62.81	53.85	45.38	32.95
99.55 8	81.87	74.03	65.86	57.65	45.01	99.49	78.70	69.98	61.58	53.25	40.22	99.23	72.55	62.92	53.88	45.59	32.78
99.59 8	81.22	73.32	66.33	58.12	44.95	99.51	78.57	69.44	61.64	53.40	40.56	99.40	72.61	62.24	53.67	45.66	32.77
					PdF	PF In	dex w	ith Sei	vice I	Level (Constra	aint					
	Pri	ce = {8	80, 80,	80}			Pri	ce = {5	55, 55,	55}			Pri	ce = {4	40, 40,	40}	
91.87 7	76.42	68.93	62.27	55.24	41.81	68.41	58.10	53.62	49.08	43.80	34.83	56.81	47.13	44.23	41.90	36.89	29.64
99.55 7	77.20	69.97	62.31	55.20	42.39	74.00	62.23	57.08	52.29	46.71	37.03	55.39	41.04	39.10	40.27	32.21	30.08
94.94 7	78.49	68.85	62.07	54.51	41.78	74.27	63.06	58.49	49.09	47.96	38.01	59.44	53.85	46.89	43.92	39.29	32.25
96.14 7	79.09	69.77	62.66	55.17	42.06	73.29	60.48	55.83	49.98	44.13	35.18	54.92	46.12	44.42	39.91	38.16	29.48
93.35 7	93.35 76.10 68.89 61.29 54.78 42.69						63.01	57.57	50.98	45.13	35.91	57.16	48.86	44.83	42.15	38.09	33.43

Appendix C: Unmet Demand Percentage

Appendix C.1:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5, σ^2 = 10000 for each of the three products; Marginal cost of capacity = 10 for any type of plant; ρ = {0.99, 0.5, 0.25, 0, -0.25, -0.5}

→ Horizontally {Unmet demand % for cases 1, 2 and 3}

Price :	= {80, 80), 80}		Price =	= {60, 80	, 100}	Price = {40, 80, 120}				
Ch	ange in	ρ		Ch	ange in	ρ	Change in p				
33.01	3.01 16.4 5.53			34.69	17.76	8.29	41.82	25.25	16.64		
32.92	16.35	5.52		34.71	17.82	8.3	41.84	25.29	16.61		
33	16.39	5.52		34.65	17.78	8.27	41.82	25.31	16.58		
32.97	16.44	5.51		34.59	17.78	8.28	41.84	25.32	16.57		
33.03	16.38	5.49		34.64	17.79	8.26	41.77	25.25	16.48		
32.94	16.39	0		34.67	17.78	0	41.9	25.28	0		

Appendix C.2:

Parameter values: Mean demand = 500, Cost = 20, Salvage value = 5, σ^2 = 10000 for each of the three products; Marginal cost of capacity = 10 for any type of plant; ρ = {0.99, 0.5, 0.25, 0, -0.25, -0.5}

Price	= {80, 80), 80}		Price :	= {55, 55	5, 55}		Price = {40, 40, 40}				
Ch	Change in p				Change in p				Change in p			
33.01	16.4 5.53			49.55	28.26	9.5		71.1	49.63	16.62		
32.92	16.35	5.52		49.61	28.29	9.47		71.03	49.55	16.61		
33	16.39	5.52		49.64	28.29	9.48		71.04	49.6	16.6		
32.97	16.44	5.51		49.62	28.22	9.47		71.06	49.57	16.59		
33.03	16.38	5.49		49.65	28.18	9.44		71.1	49.61	16.52		
32.94	16.39	0		49.53	28.24	0		71.13	49.58	0		

→ Horizontally {Unmet demand % for cases 1, 2 and 3}

Appendix C.3:

Parameter values: Mean demand = 500, Price = 80, Cost = 20, Salvage value = 5, σ^2 = 10000 for each of the three products; $\rho = \{0.99, 0.5, 0.25, 0, -0.25, -0.5\}$

→ Horizontally {Unmet demand % for cases 1, 2 and 3}

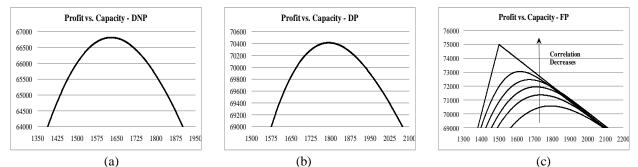
	Capacity cost = 5				Capao	city cost	= 10	Capacity cost = 15			
	Change in p				Ch	ange in	ρ	Change in p			
2	26.28	5.28 8.17 2.76			33.01	16.4	5.53	39.61	24.68	8.29	
2	26.37	8.16	2.76		32.92	16.35	5.52	39.59	24.66	8.31	
2	26.31	8.18	2.76		33	16.39	5.52	39.6	24.64	8.28	
2	26.31	8.19	2.73		32.97	16.44	5.51	39.61	24.68	8.25	
	26.3	8.22	2.73		33.03	16.38	5.49	39.59	24.68	8.25	
4	26.33 8.18 0			32.94	16.39	0	39.64	24.67	0		

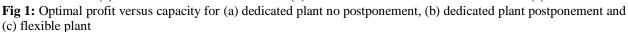
Appendix D: Incremental Profit Value

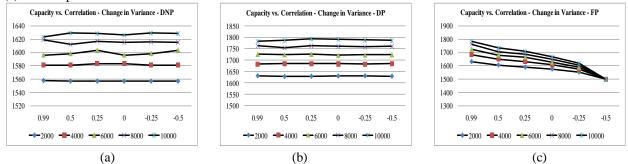
	Price = {80, 80, 80}														
	Dedica	ated Plant	t, Postpon	ement			Flexi	ble Plant,	Postpone	ment					
-0.26	-0.13	-0.22	-0.17	-0.21	-0.25	-0.08	0.03	-0.06	-0.05	-0.14	0.00				
-0.12	-0.33	-0.44	-0.40	-0.23	-0.20	-0.12	-0.14	-0.27	-0.17	-0.07	0.00				
-0.51	-0.12	-0.36	-0.16	-0.19	-0.31	-0.33	0.00	-0.08	0.06	0.06	0.00				
-0.23	-0.57	-0.25	-0.36	-0.33	-0.33	-0.08	-0.41	0.02	-0.15	-0.15	0.00				
0.01	-0.93	-0.59	-0.12	-0.43	0.33	0.33	-0.60	-0.33	0.26	-0.10	0.00				
	Price = {55, 55, 55}														
	Dedica	ated Plant	t, Postpon	ement		Flexible Plant, Postponement									
-1.43	-1.40	-1.40	-1.53	-1.45	-1.40	0.02	0.05	0.04	-0.04	0.01	0.00				
-1.91	-1.77	-1.31	-1.50	-1.59	-1.71	-0.26	-0.08	0.37	0.17	0.09	0.00				
-1.93	-2.21	-2.11	-1.80	-1.97	-1.99	0.05	-0.31	-0.12	0.10	-0.06	0.00				
-2.45	-2.33	-2.14	-2.19	-2.20	-2.41	-0.14	0.03	0.16	0.12	0.17	0.00				
-2.39	-2.49	-2.32	-2.74	-3.02	-2.76	0.22	0.03	0.17	0.01	-0.19	0.00				
					Price = {4	0, 40, 40}									
	Dedica	ated Plant	t, Postpon	ement			Flexi	ble Plant,	Postpone	ment					
-6.04	-6.08	-5.92	-6.06	-6.01	-6.02	-1.66	-1.04	-1.23	-1.27	-0.43	0.00				
-9.60	-9.69	-9.93	-8.63	-9.75	-9.19	-2.00	-1.37	-1.87	-1.00	-1.21	-2.00				
-10.89	-10.20	-11.12	-10.83	-10.91	-11.17	-2.85	-2.00	-1.64	-1.54	-1.01	0.00				
-13.03	-13.25	-13.50	-13.12	-13.01	-12.88	-2.59	-2.08	-2.60	-1.38	-0.83	0.00				
-14.25	-14.83	-14.98	-14.86	-14.04	-13.99	-3.25	-3.34	-2.19	-2.36	-1.73	0.00				

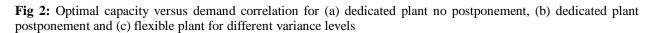
C	hange in I	Price Diff	erential =	: {0, 40, 80	0}	Change in Capacity Cost = {5, 10, 15}								
	Dedicated Plant,			exible Pla	/		licated Pla			exible Pla	/			
Po	Postponement			stponeme	ent	Po	stponeme	nt	Po	Postponement				
0.01	-0.49	-1.09	0.33	-0.01	-0.64	0.35	0.01	-1.48	0.35	0.33	0.38			
-0.93	-0.24	-1.31	-0.60	0.30	-0.81	0.71	-0.93	-2.10	0.73	-0.60	-0.48			
-0.59	-0.62	-0.90	-0.33	-0.13	-0.31	-0.45	-0.59	-2.05	-0.43	-0.33	-0.27			
-0.12	-0.74	-0.65	0.26	-0.22	0.08	0.16	-0.12	-2.00	0.11	0.26	-0.16			
-0.43	-0.32	-1.17	-0.10	0.15	-0.47	-0.03	-0.43	-1.64	-0.04	-0.10	0.08			
0.33	-0.58	-0.76	0.00	-0.07	0.05	-0.05	0.33	-1.81	0.00	0.00	0.00			

Appendix E: Graphs









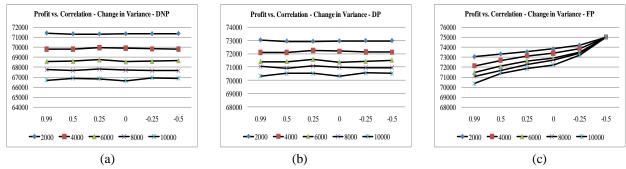


Fig 3: Optimal profit versus demand correlation for (a) dedicated plant no postponement, (b) dedicated plant postponement and (c) flexible plant for different variance levels

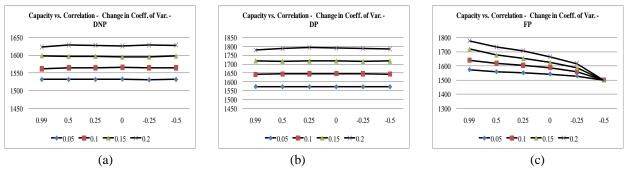


Fig 4: Optimal capacity versus demand correlation for (a) dedicated plant no postponement, (b) dedicated plant postponement and (c) flexible plant for different coefficient of variation levels

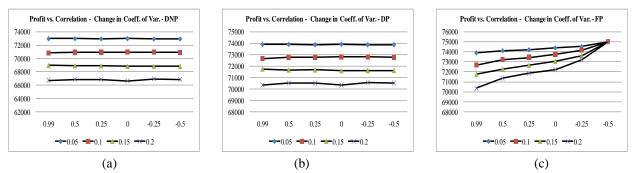
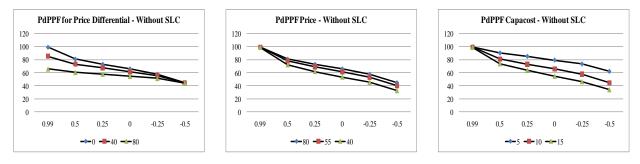
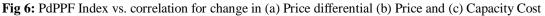


Fig 5: Optimal profit versus demand correlation for (a) dedicated plant no postponement, (b) dedicated plant postponement and (c) flexible plant for different coefficient of variation levels

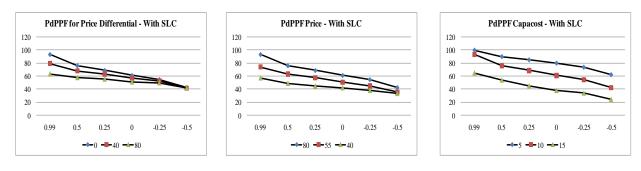


(b)

(c)



(a)



(a) (b) (c) **Fig 7:** Effect of Service Level Constraint (SLC) on PdPPF Index vs. correlation for change in (a) Price differential (b) Price and (c) Capacity Cost